

# Verification of NMME Extreme Seasonal Temperatures

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# DATA

- NMME hindcast:
  - 12 x 1-month lead seasonal hindcasts  
(e.g., November 1st initialization predictions of DJF Temperature)
  - 29 year hindcasts, 1982 to 2010
- *GHCN+CAMS 2-m temperature observations*
- Considering the skill and calibration of probability forecasts for bottom and top deciles (i.e. 10<sup>th</sup> and 90<sup>th</sup> percentile forecasts)
- ... and calibration of ensemble probabilities for above and below normal terciles simultaneously

# METHODS

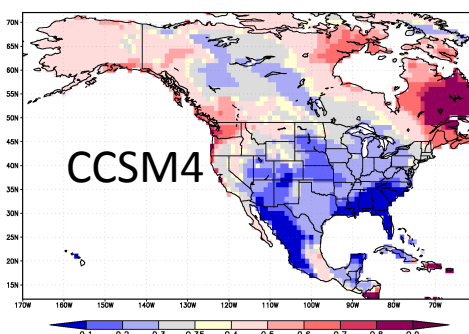
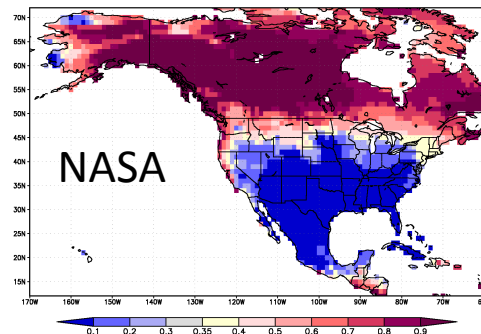
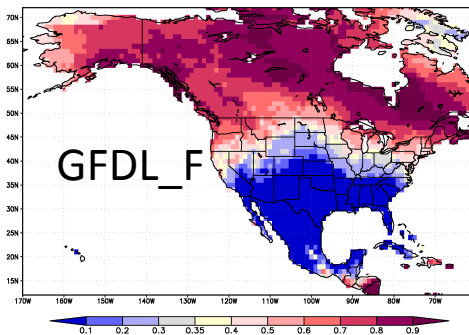
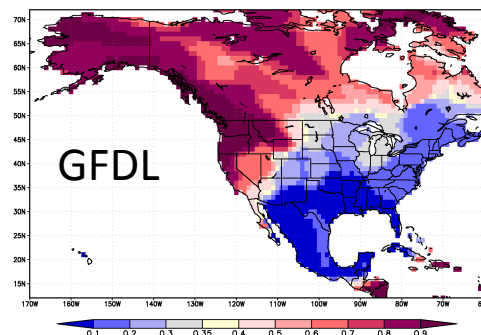
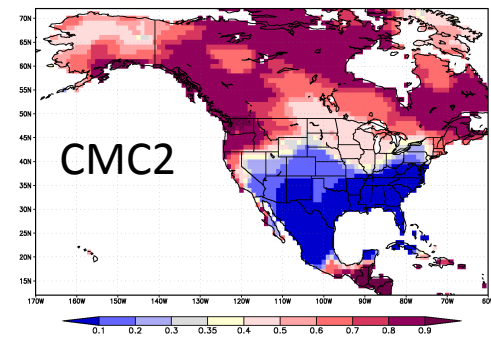
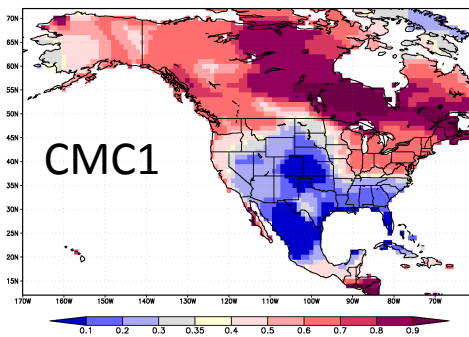
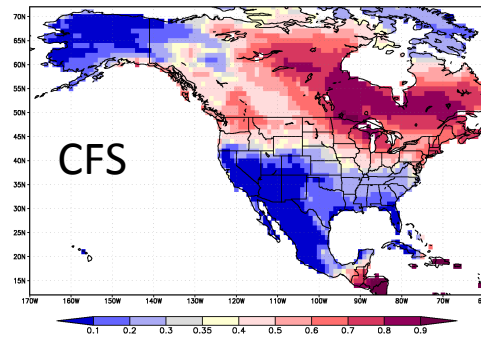
## 1. **Count:** Probabilities based on percent count of ensemble members

- Model climatological mean removed
- Variance of members corrected
- Terciles ( $\pm 0.43$  standard deviations) and deciles ( $\pm 1.28$  standard deviations) derived from normal distribution fit to hindcasts
- Cross-validated (Leave 1-year out) calculation of model mean and standard deviations

## 2. **Ensemble Regression or EReg** probability calibration methodology (Unger et al 2009).

- Removes systematic bias and calibrates probability based on hindcast correlation
- Cross-validated regression (Leave 1-year out) calculation of all regression parameters and probabilities
- Fit Gaussian distribution around each ensemble member

# Multi-model ensemble **count** forecasts



**Hindcasts of the 1997-98 DJF  
climate of North America**



# Ensemble Regression

- 1) Derive a regression equation for the least-squared error solution between the ensemble members and the observation, based on the ensemble mean and observations.

$$F^*_{(m)} = aF_{(m)} + b$$

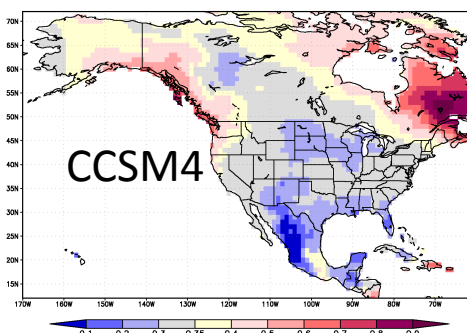
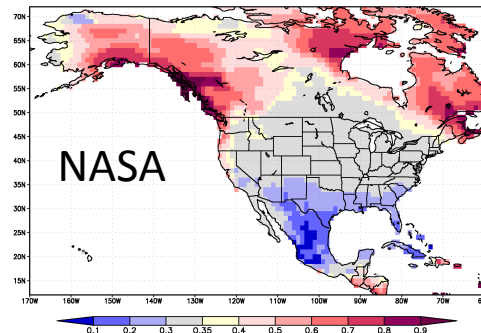
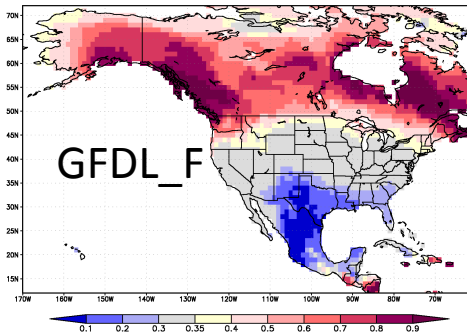
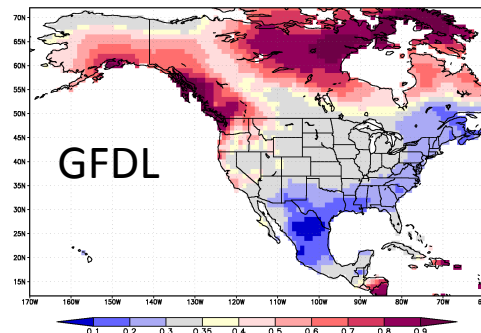
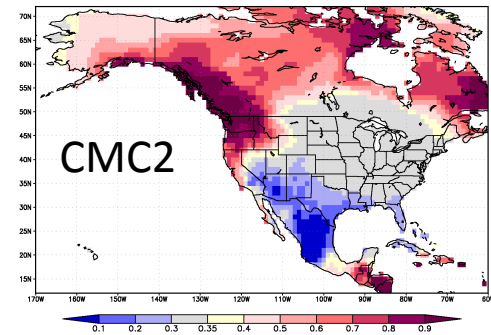
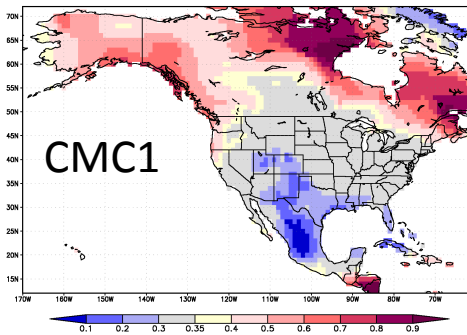
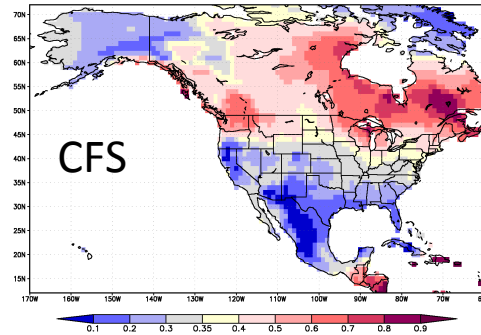
- 2) Expected residual error distribution of each member based on expected MSE of ensemble mean minus average ensemble spread.

$$\sigma_{\varepsilon}^2 = [MSE] = \sigma_{ens}^2 + \varepsilon^2$$

$$\varepsilon^2 = \sigma_{\varepsilon}^2 - \sigma_{ens}^2$$

$$[MSE] = \sigma_{obs}^2 (1 - R_m^2)$$

# Multi-model ensemble **EReg** calibrated forecasts

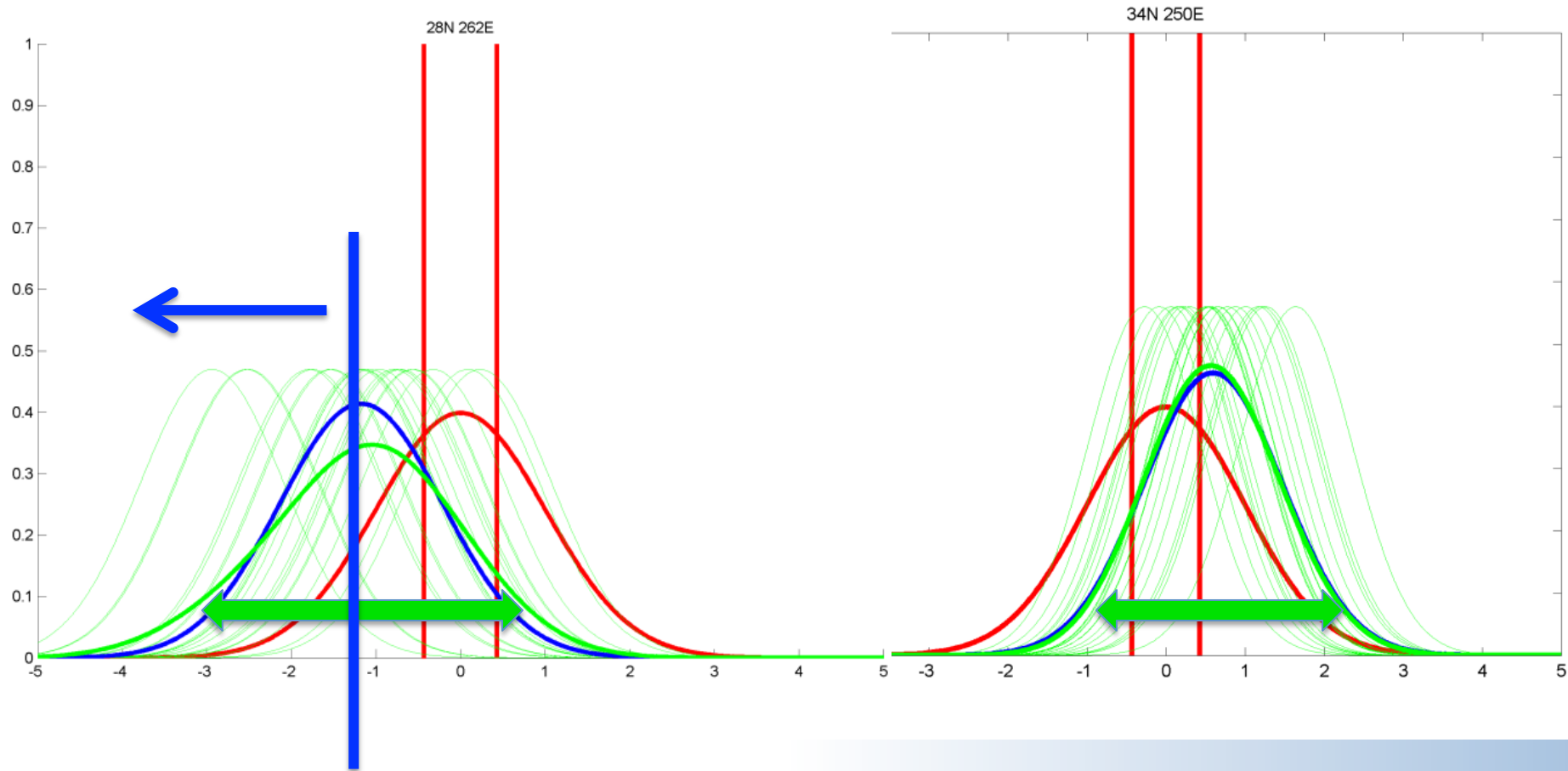


**Calibration makes maps  
equally reliable and nearly  
equally skillful , but not the  
same.**

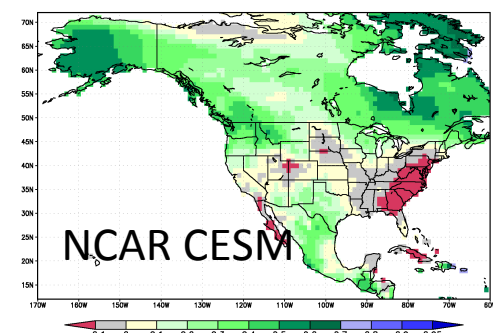
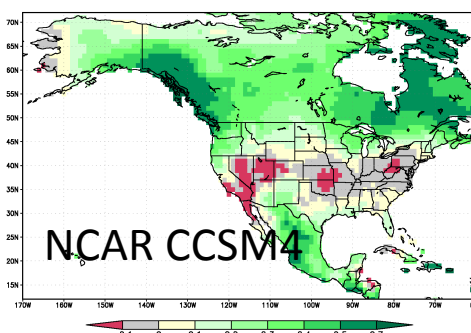
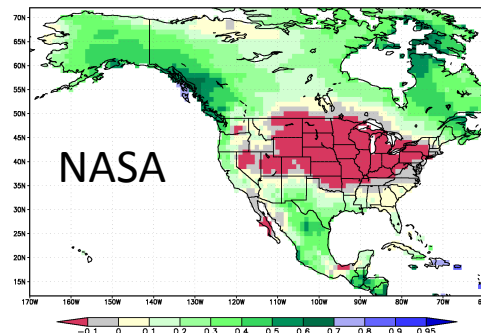
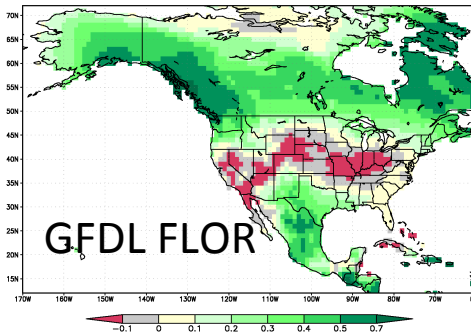
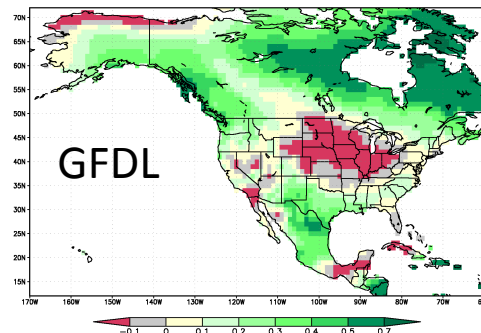
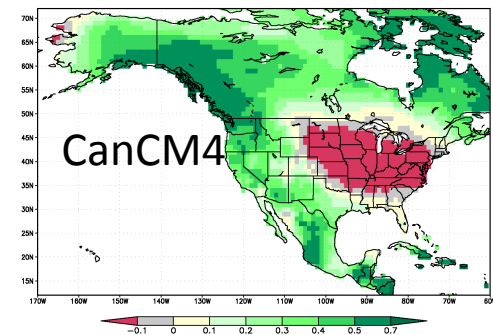
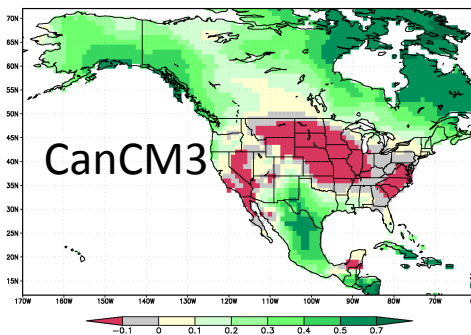
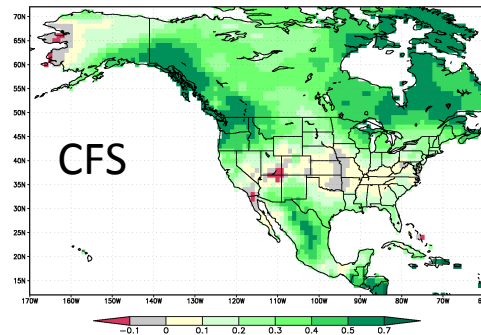
# Seasonal Forecasting of Extremes & MME

- Low predictability & extremes due to signal+noise
- Some individual models have 10 ensemble members ->
  - Poor resolution of tails of distribution (extremes)
- Probabilistic Outlooks for categorical forecasts
  - 3-category tercile forecasts (Above and Below normal)
  - Probability of extremes should be consistent with tercile forecasts
- Correlation used to adjust spread of model
  - Realistic probabilities representing hindcast skill
  - Improves **reliability**, while maintaining resolution of events
  - **Brier Skill Score** as metric (Resolution + Reliability)
- Multi-Model Ensembles
  - Many ensemble members reducing noise and canceling errors
  - **Better resolution of probability distribution, including tails (i.e. extremes)**

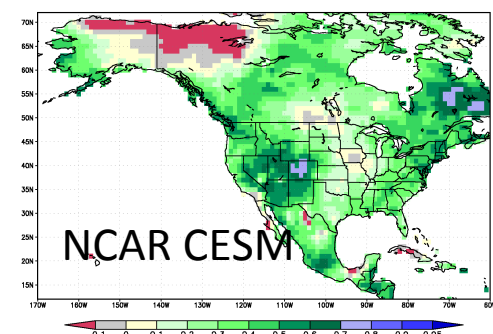
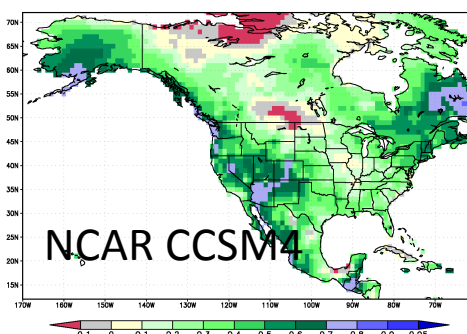
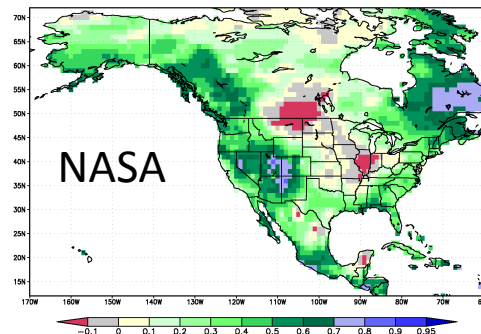
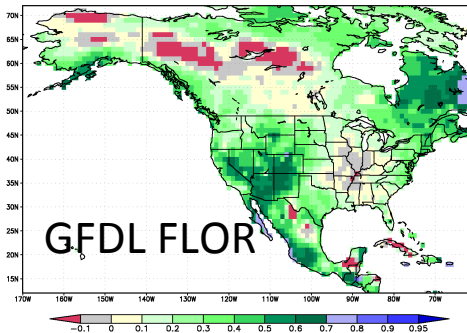
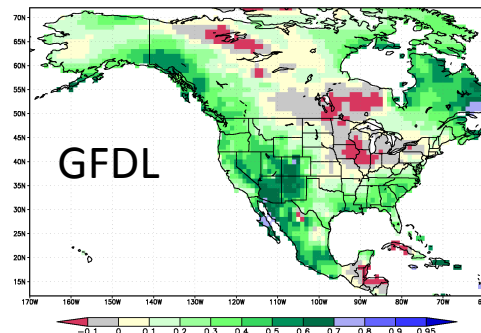
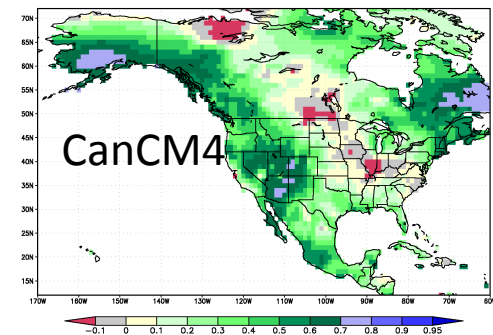
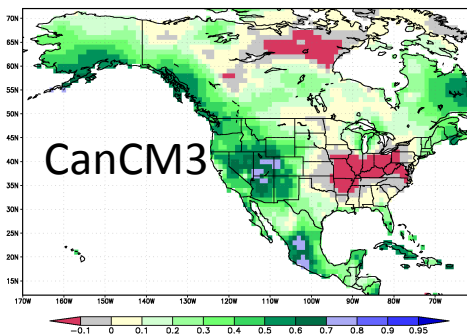
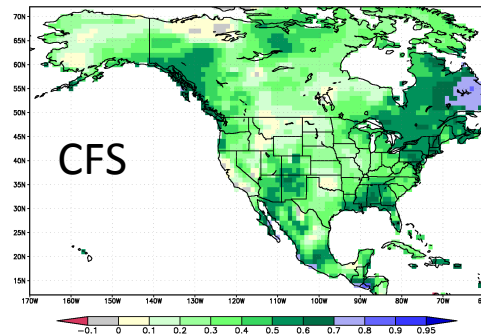
# Variations in the mean and spread of ensemble members: Probability of extremes $< -1.28$ sigma, or $> +1.28$ sigma



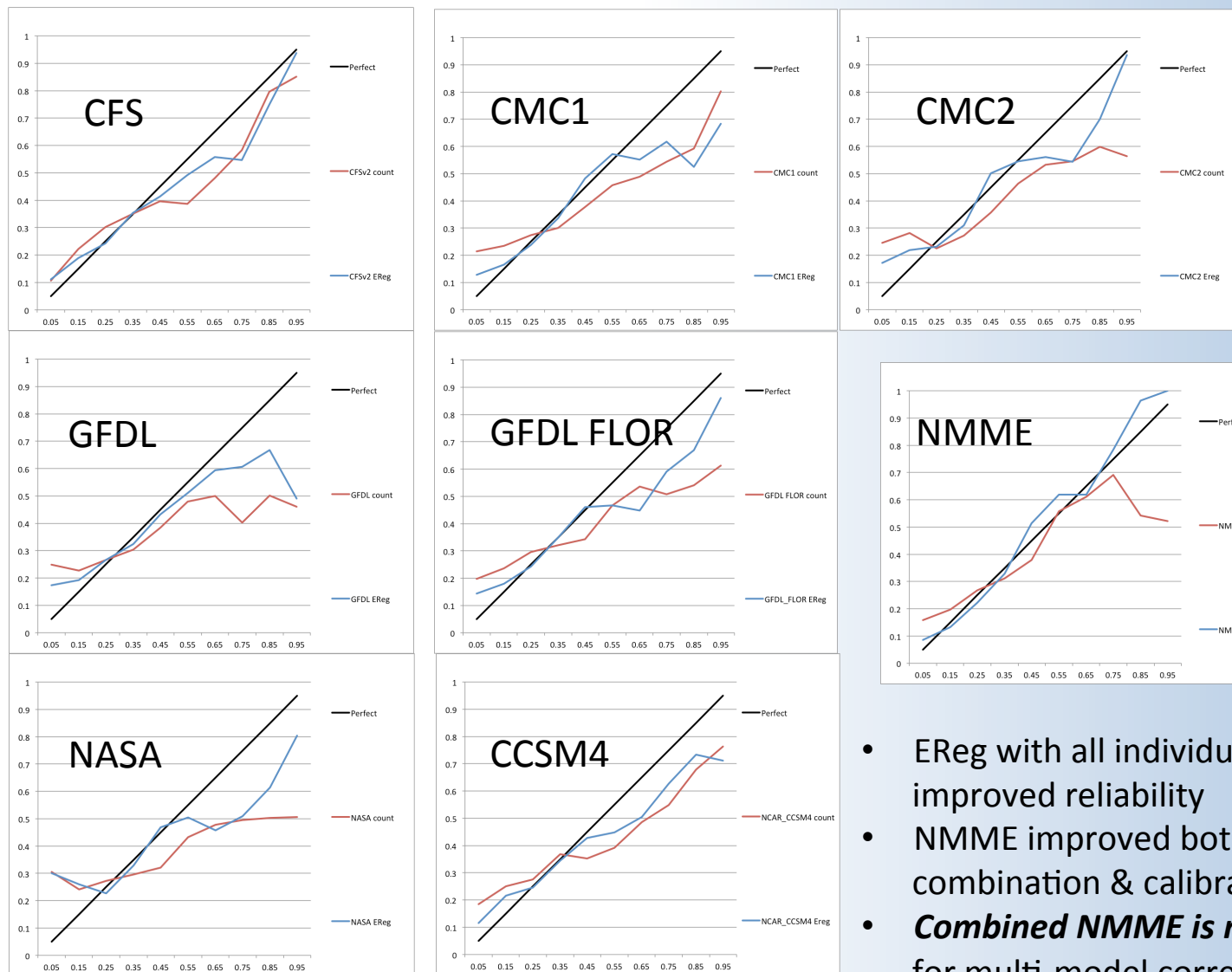
# Cross-validated DJF Correlation



# Cross-validated JJA Correlation



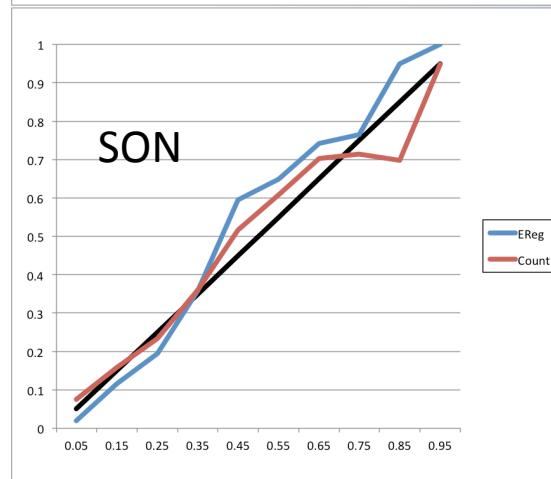
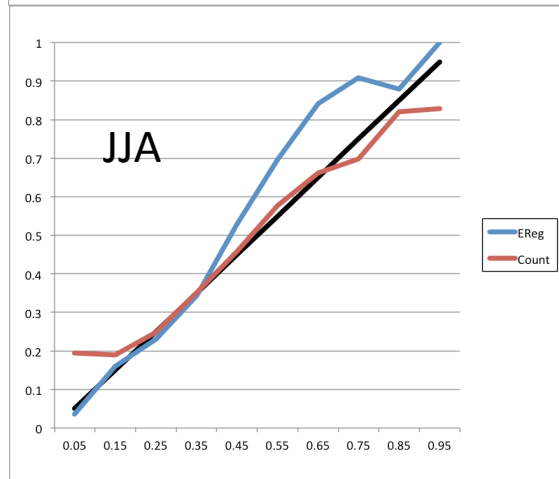
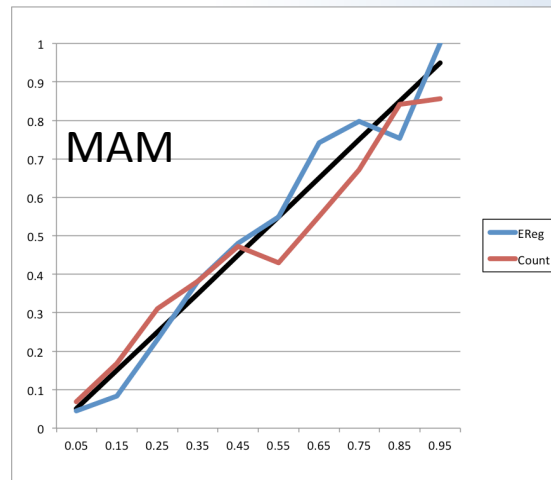
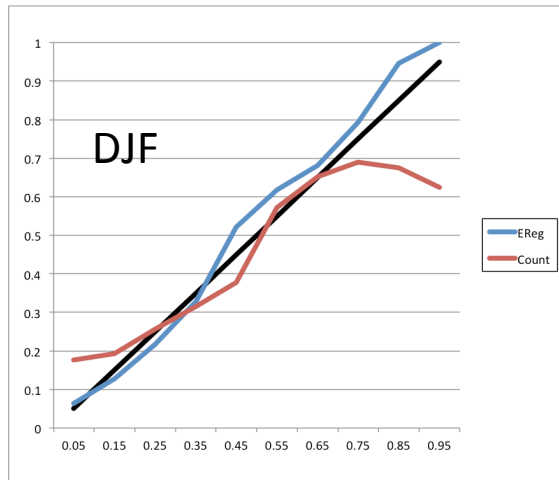
# Each individual model's PDF is calibrated to improve reliability (DJF Terciles shown)



- EReg with all individual models -> improved reliability
- NMME improved both by MME combination & calibration
- ***Combined NMME is not adjusted*** for multi-model correlation



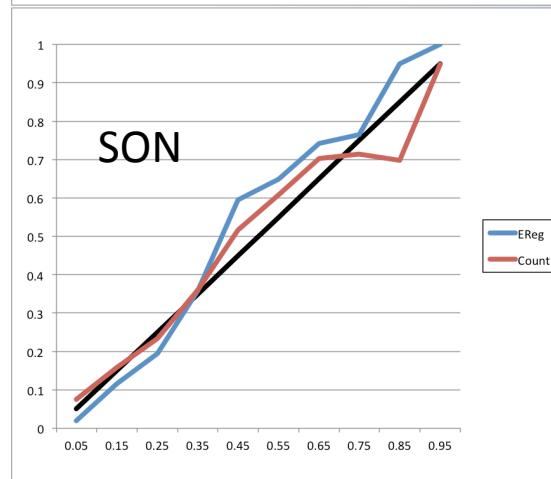
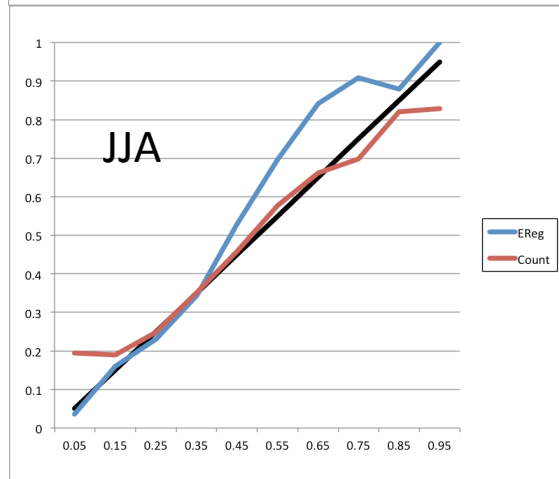
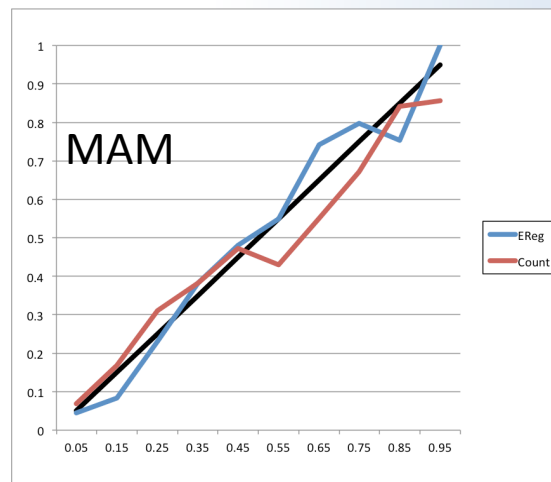
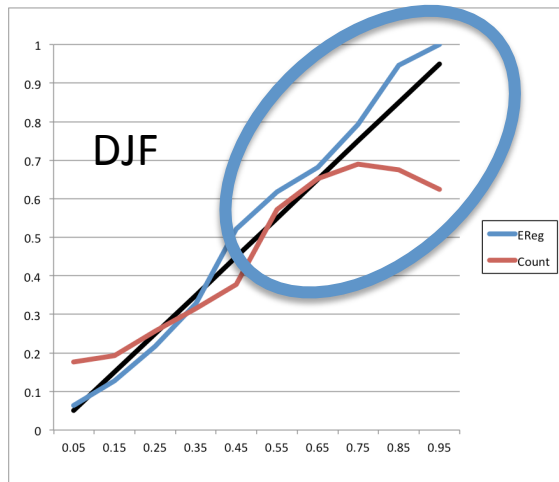
# Combined NMME Count & EReg Tercile Probability Reliability



- Calibration improves DJF & MAM reliability
- Combined EReg NMME JJA & SON probabilities under-confident
- **Probabilities of combined NMME somewhat reliable. Combining calibrated models -> under-confident total probability**

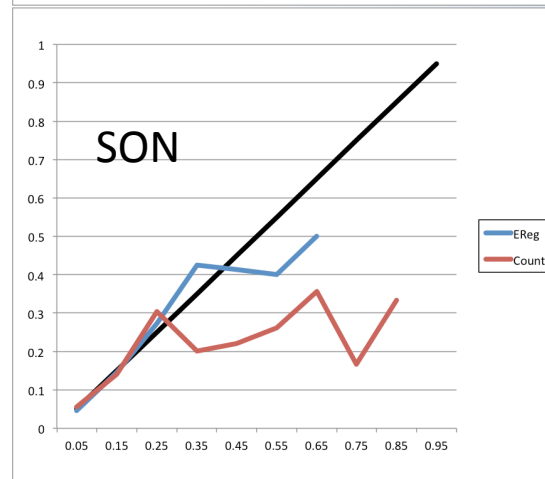
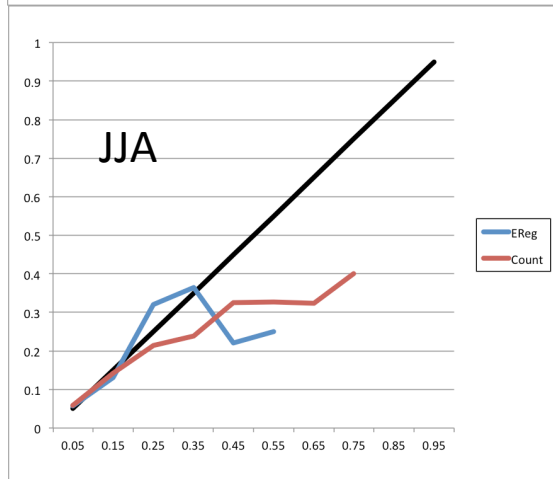
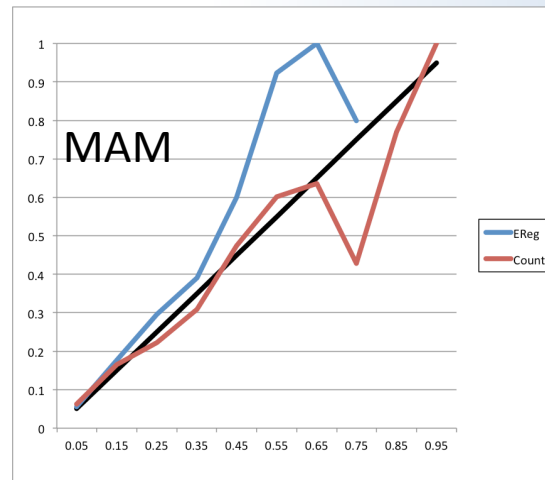
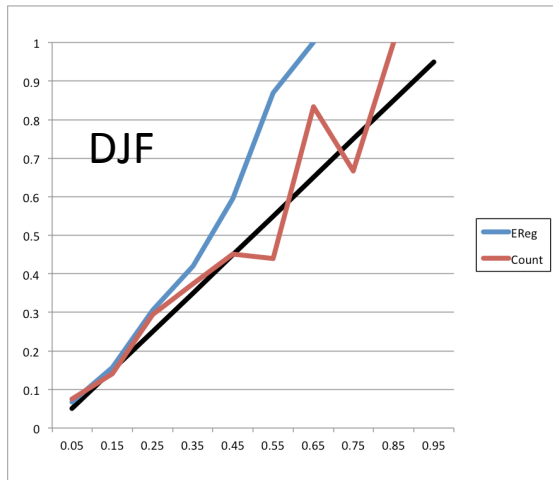


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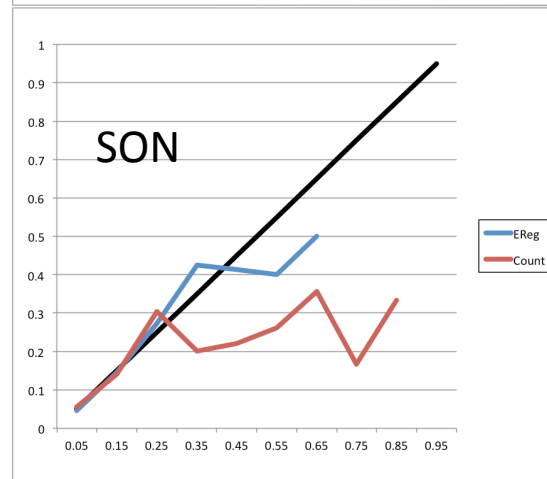
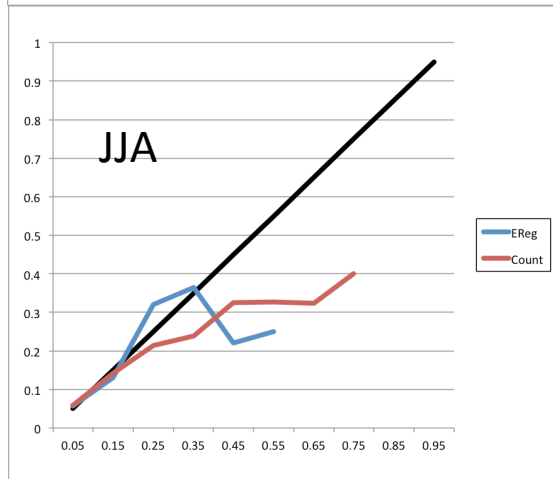
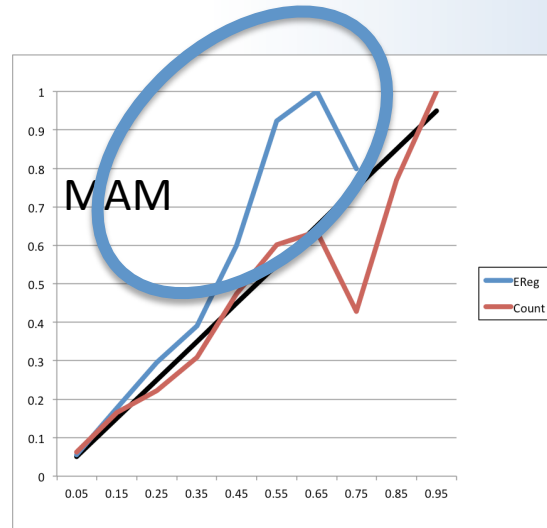
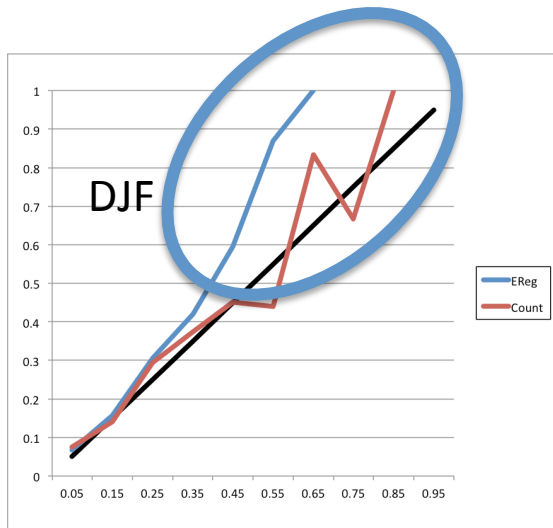
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# Combined NMME **Count** & **EReg** Extreme Probability Reliability



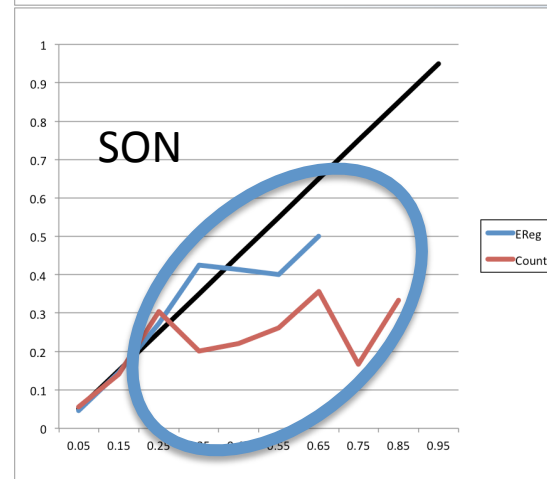
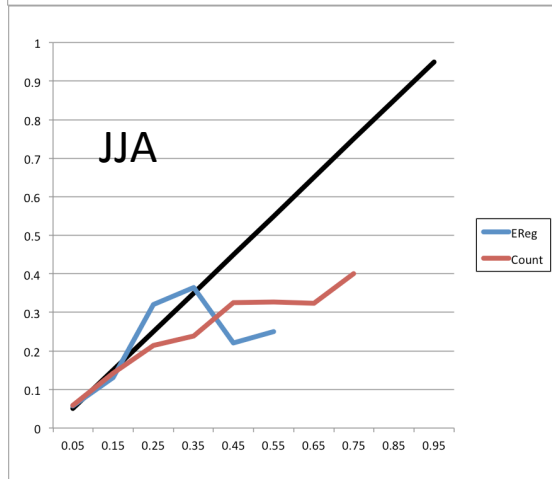
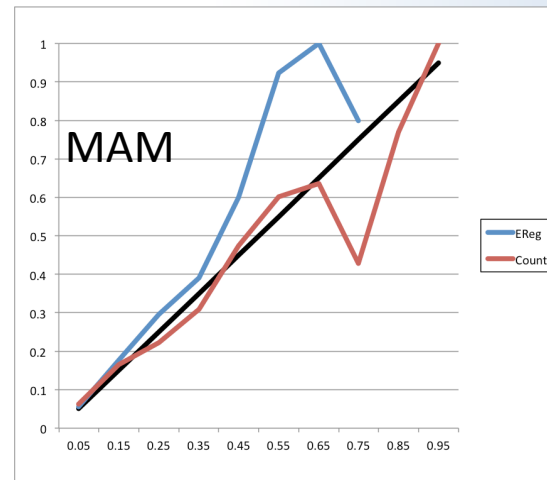
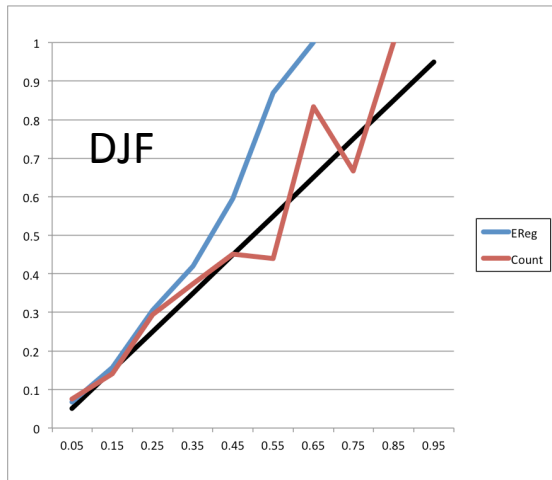
- Calibration of PDF produces under-confident forecasts in DJF & MAM
- Better reliability in some seasons

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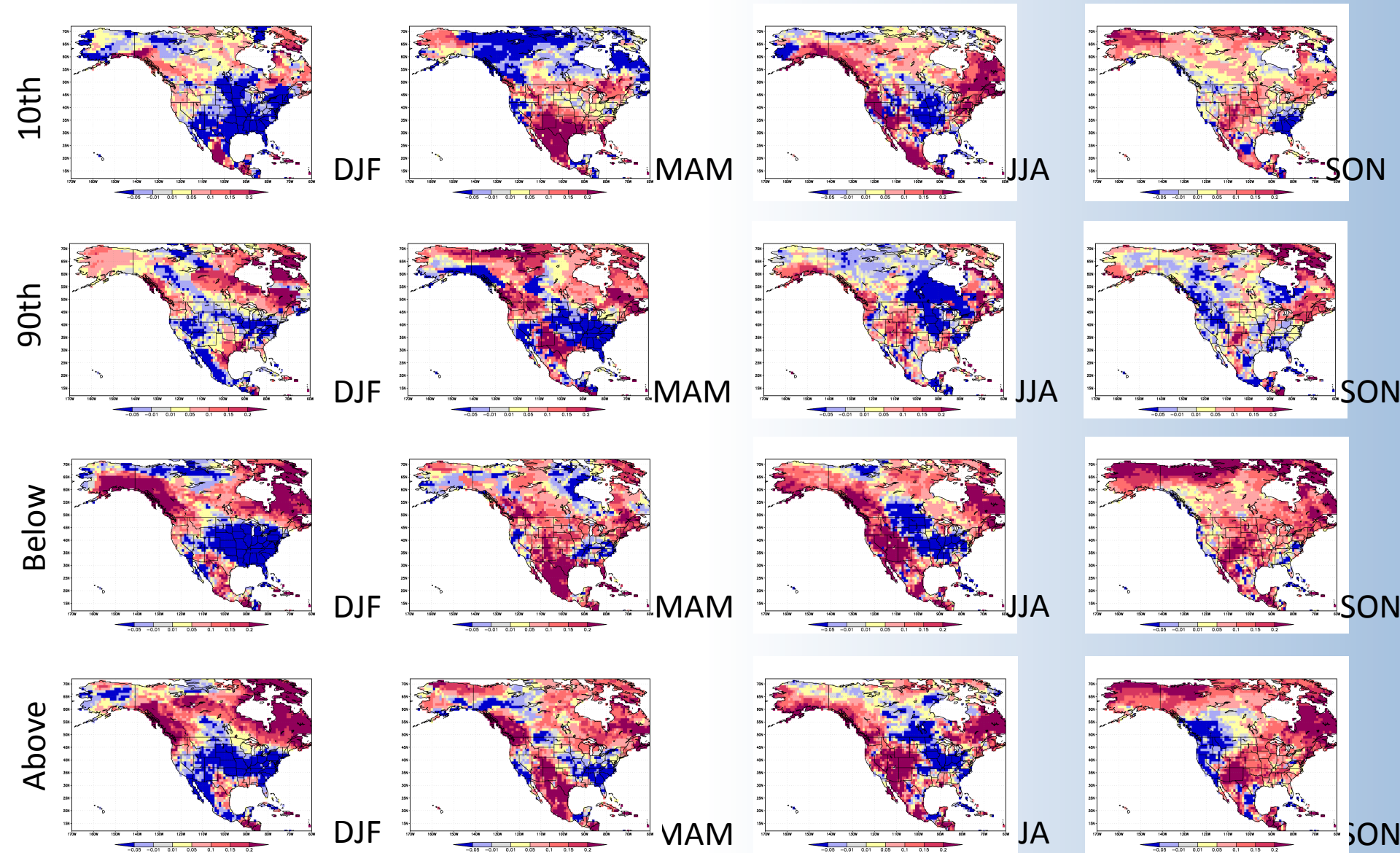
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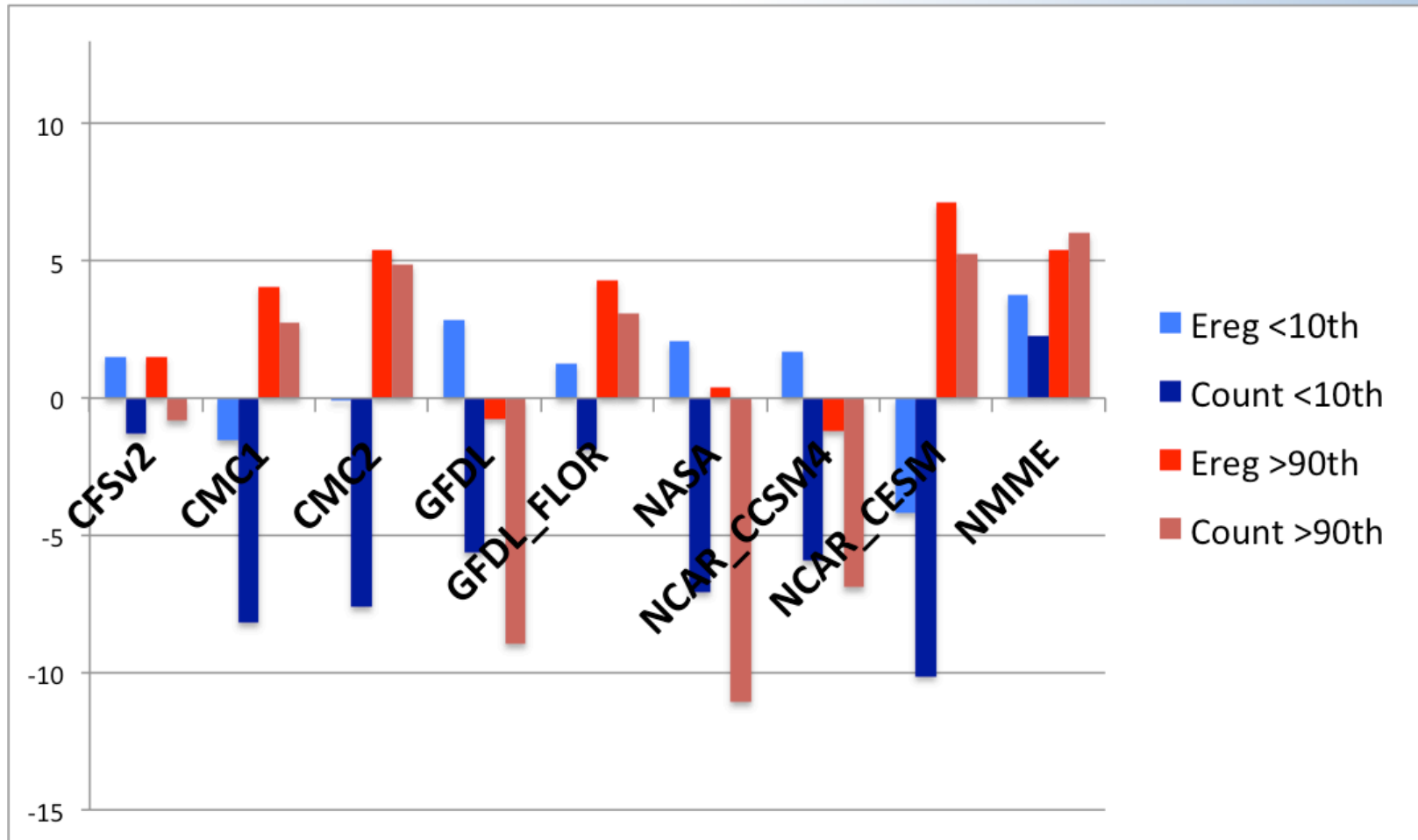
# Four seasons 10<sup>th</sup> (top) and 90<sup>th</sup> (2<sup>nd</sup> row) percentile **count** hindcast Brier Skill Scores compared to below and above normal tercile BSS





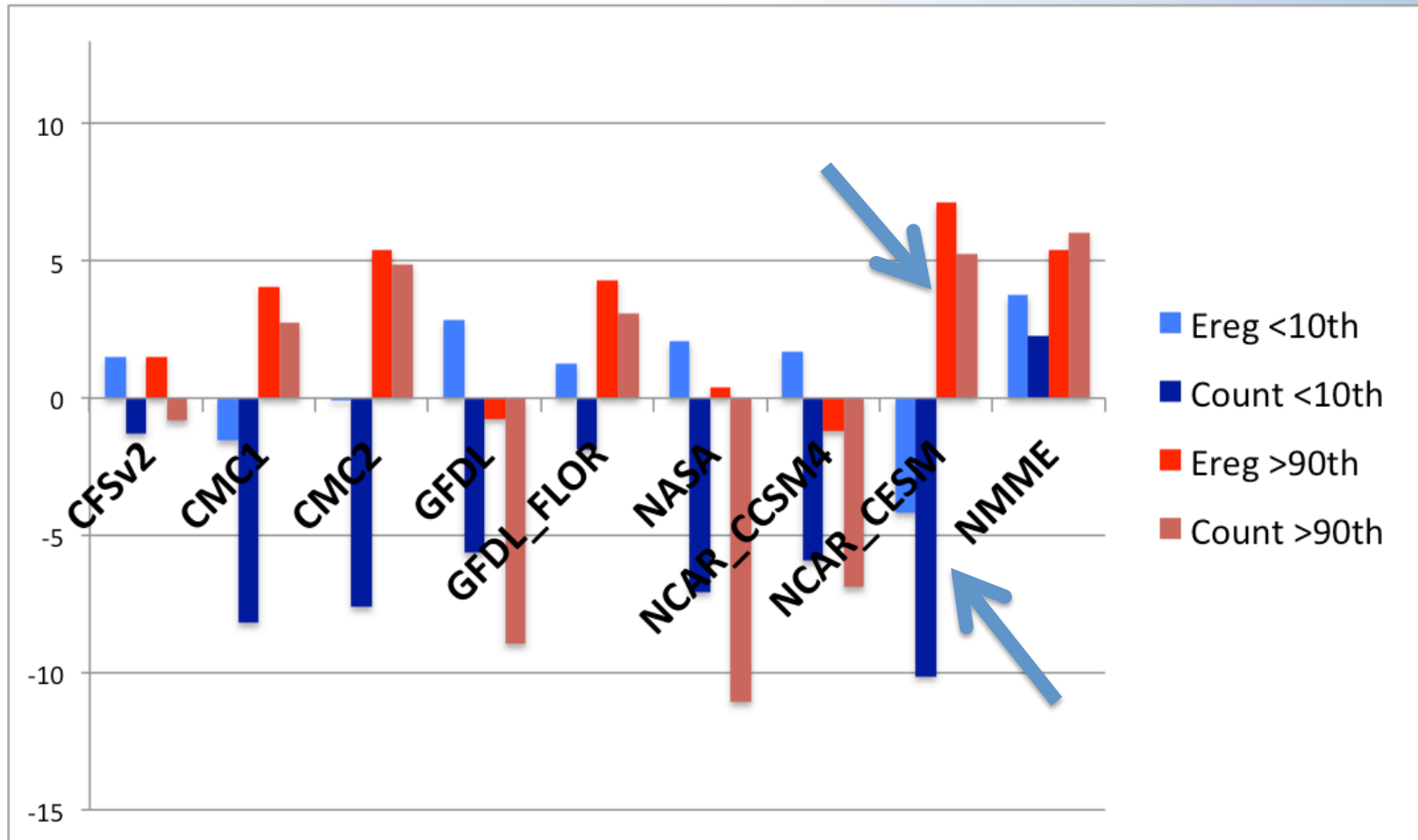
## North America average Brier Skill Scores

**DJF** climatological 10<sup>th</sup> (blue) and 90<sup>th</sup> (red) percentile hindcasts  
8 models and the combined NMME, bias corrected **Count** (bright) &  
Ensemble Regression or **Ereg** calibrated (darker colors)



## North America average Brier Skill Scores

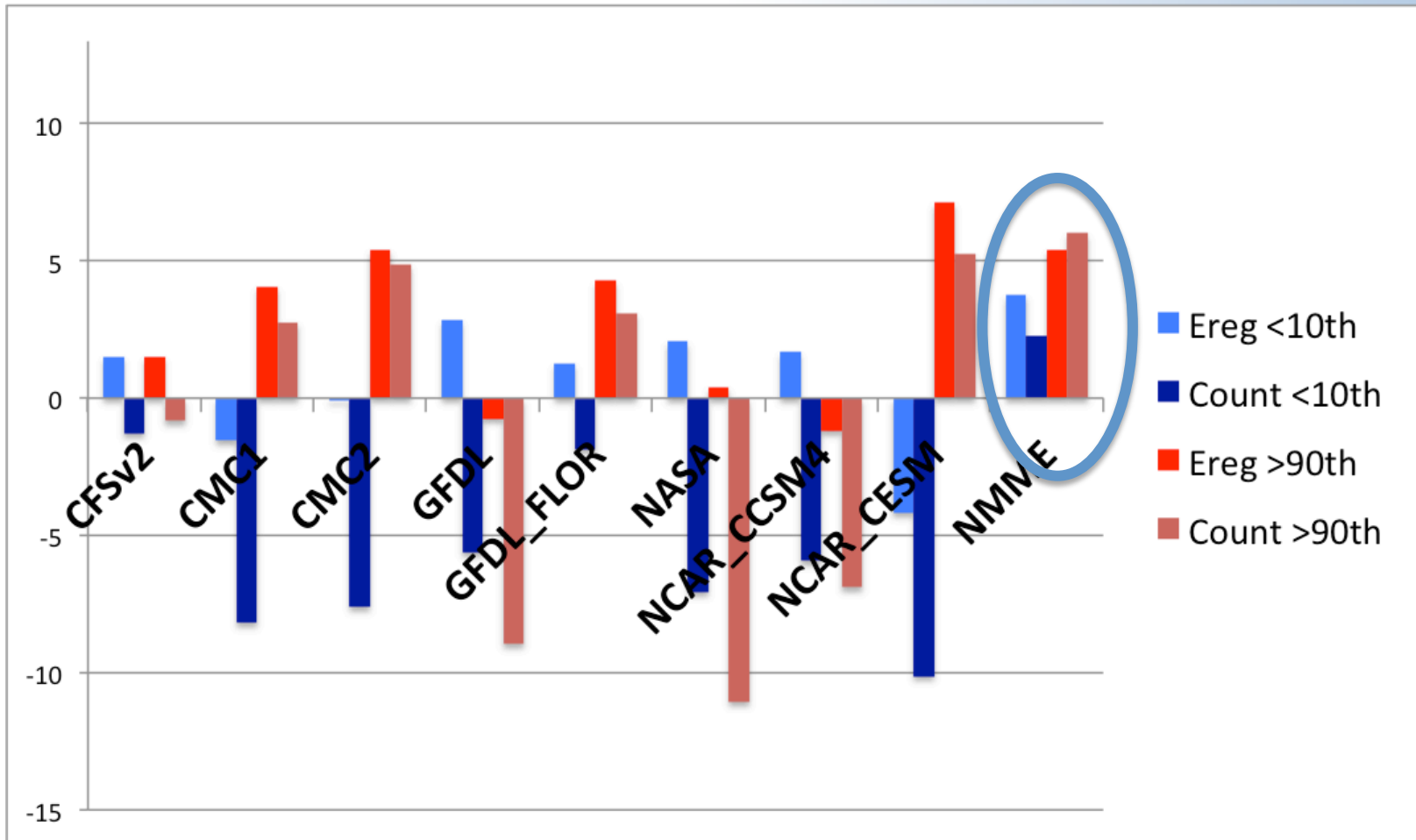
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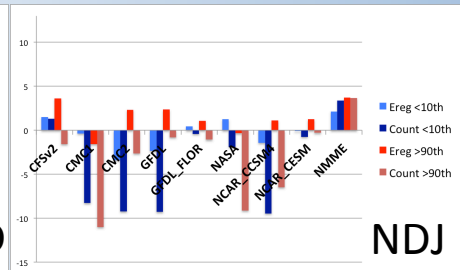
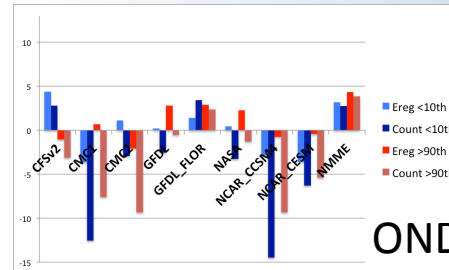
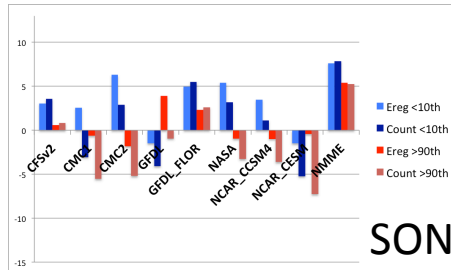
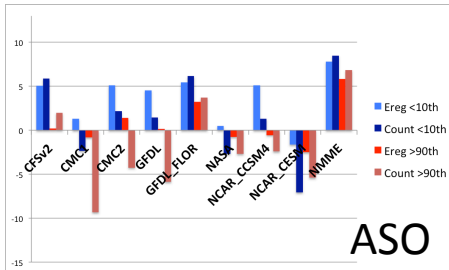
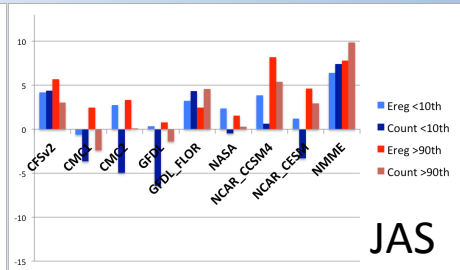
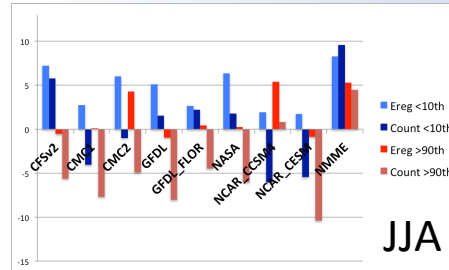
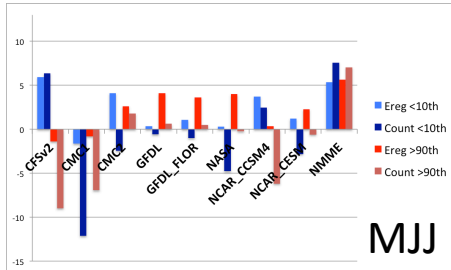
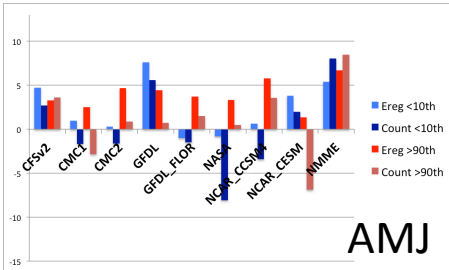
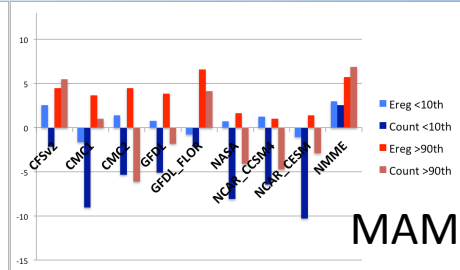
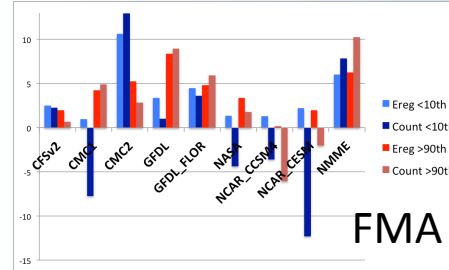
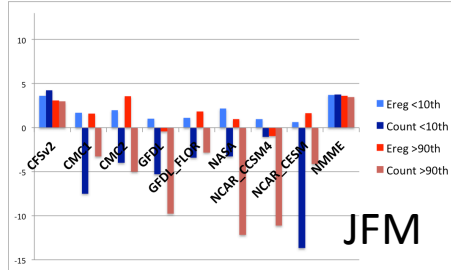
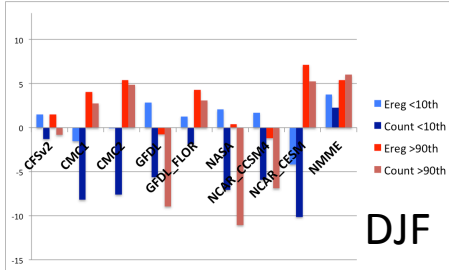
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## North America average Brier Skill Scores

All 12 seasons 10<sup>th</sup> (blue) and 90<sup>th</sup> (red) percentile hindcasts

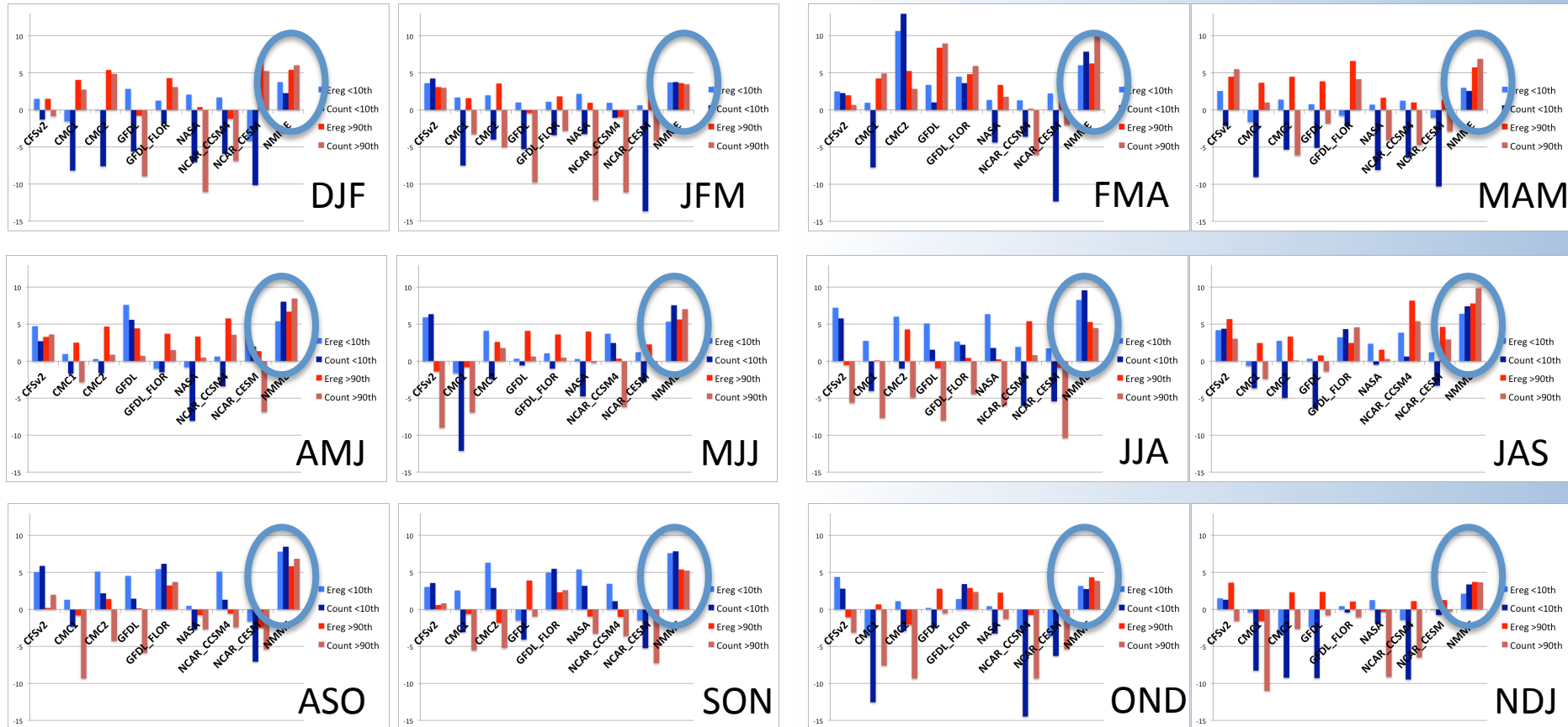
\*Combined NMME (far right)



# North America average Brier Skill Scores

All 12 seasons 10<sup>th</sup> (blue) and 90<sup>th</sup> (red) percentile hindcasts

\*Combined NMME (far right)



Combined NMME has positive skill in all seasons.

Not always the greatest skill, but consistently near the best model

## Results & Conclusions

- Skill of tercile probabilities of seasonal forecasts implies skill of extremes
  - Skill in extreme forecasts in same regions as tercile forecasts
- Patterns of skill of 10<sup>th</sup> and 90<sup>th</sup> percentile forecasts similar; Differences appear related to decadal trends
- While individual ensemble models often have negative skill when forecasting extremes, on average over North America, **the combined NMME is found to have skill when forecasting extremes in all seasons**
- Regression calibration successfully removes areas of negative skill; however, combined NMME forecasts are sometimes under-confident

Thanks

